

# Core C++ 2024

# C++ ♥ Python

Alex Dathskovsky



C++  
♥  
python



# About Me:

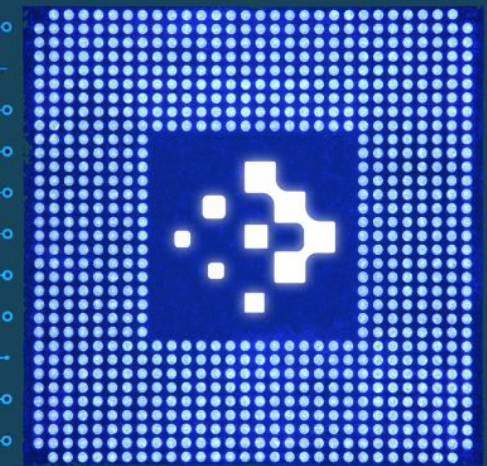


**[alex.dathskovsky@speedata.io](mailto:alex.dathskovsky@speedata.io)**

**[www.linkedin.com/in/alexdatahskovsky](https://www.linkedin.com/in/alexdatahskovsky)**

**[www.cppnext.com](http://www.cppnext.com)**

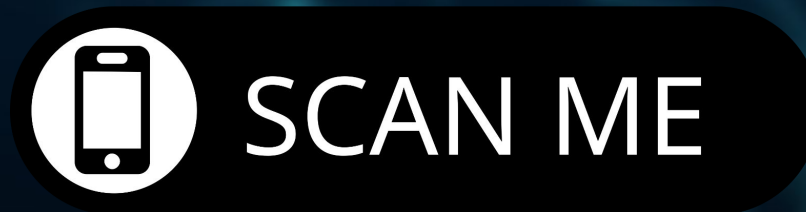
**<https://www.youtube.com/@cppnext-alex>**





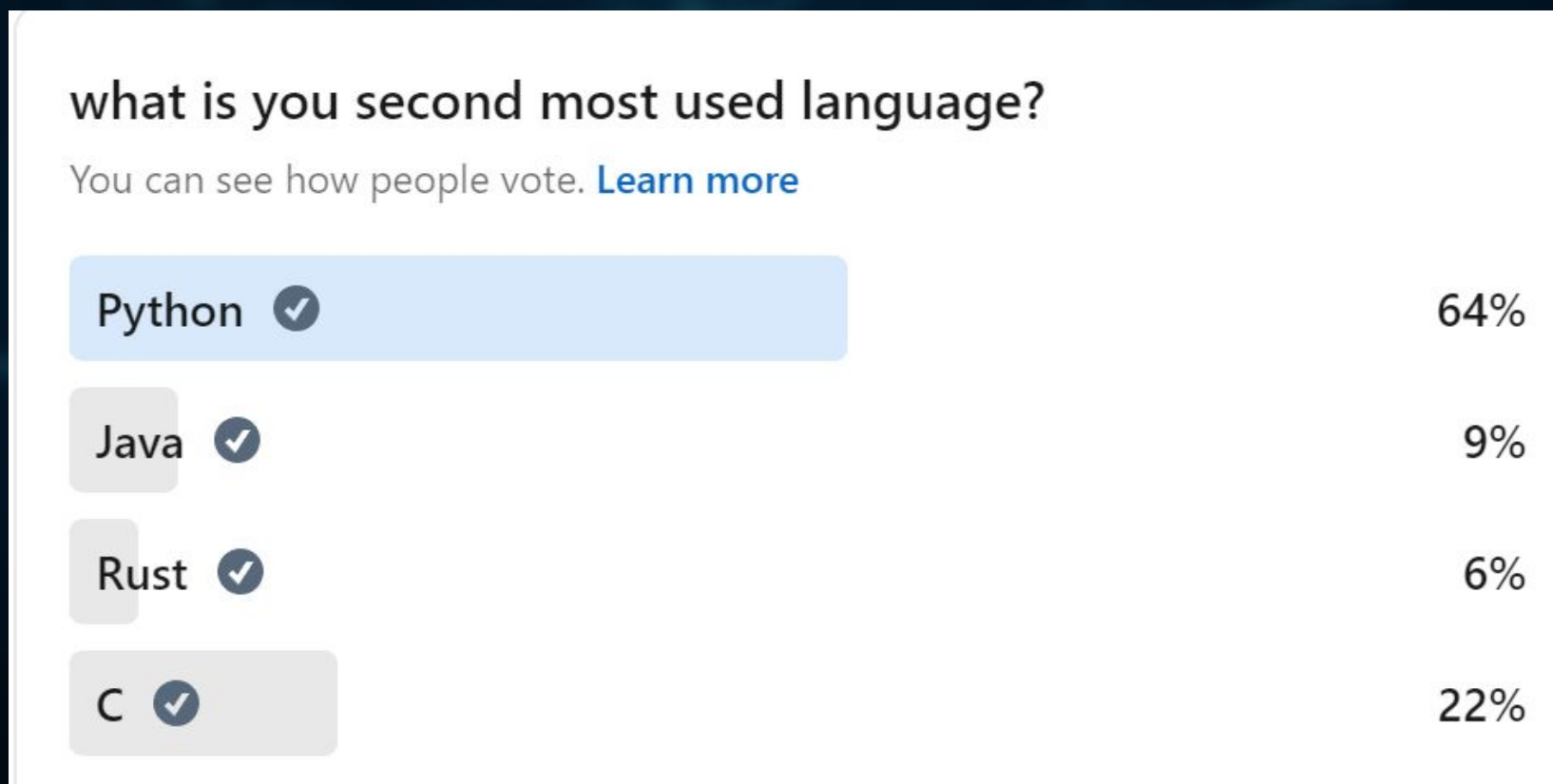
# Survey Time

# CoreC++2024 Survey



# Why Python????

- My survey



# Why Python????

- C++ standard survey

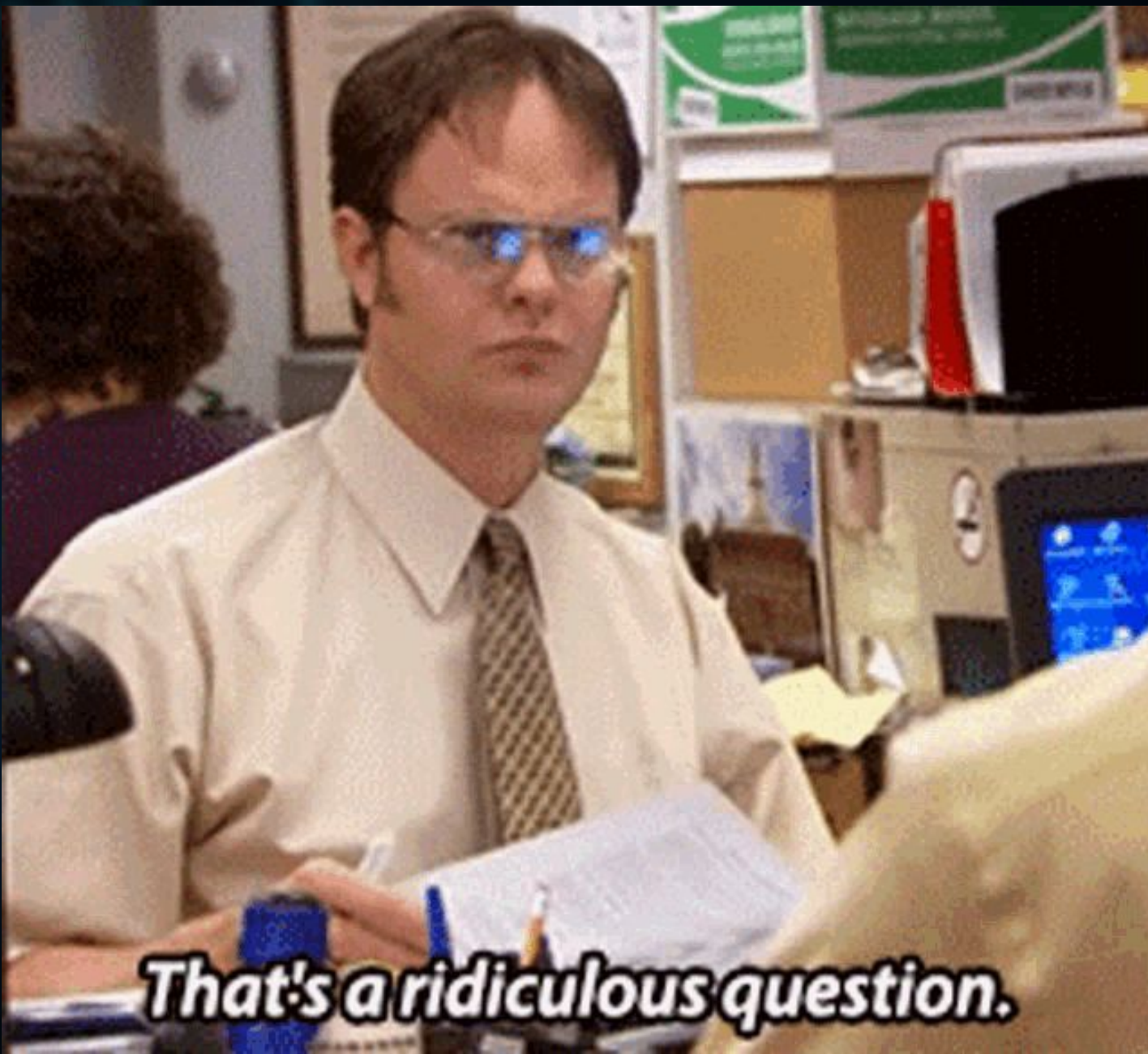
Q16 Besides C++, what programming languages/environments do you use in your current and recent projects? (select all that apply)

Answered: 1,225 Skipped: 40

Python	72.82%	892
C	53.06%	650
JavaScript	27.43%	336
C#	23.76%	291

# Why C++ ?





***That's a ridiculous question.***

# Why Python????

# Why Python????

- Faster Development
- Simpler Interface/usage
- Interpreted language
  - Easy way to research algorithms



# Similarities Between C++ and Python

# Some Interfaces similar to python in C++

Python:

```
1 print(1, 2, this is 3 {1+2})
```

# Some Interfaces similar to python in C++

C++:

```
4  int main(){
5  |   fmt::pr(1, 2) this is 3   tuple(1,2), 3);
6  }
```



# Some Interfaces similar to python in C++

Python:

```
1 for i in [1, 2, 3]:  
2     print(i)  
3
```

# Some Interfaces similar to python in C++

C++:

```
4  int main(){
5      for (auto i : std::array{1, 2, 3}){
6          |   fmt::print("{}\n", i);
7          |   }
8      }
```

# Some Interfaces similar to python in C++

Python:

```
1 def foo(a, b):  
2     return a, b, a+b  
3  
4 a, (1, 2, 3), 4  
5  
6 print(f"{foo(1, 2)}, {a+c}")
```



# Some Interfaces similar to python in C++

C++:

```
1  #include <tuple>
2  #include <fmt/ranges.h>
3
4  auto foo(auto x, auto y){
5      |   return std::tuple{x, y, x+y};
6  }      (1, 2, 3), 4
7
8  int main(){
9      |   auto [a, b, c] = foo(1, 2);
10     |   fmt::print("{} , {}", foo(1, 2), a+c);
11     |   }
```

# Some Interfaces similar to python in C++

Python:

```
def foo(a, b):  
    return a, b, a+b  
  
_, _, c = foo(1, 2)
```

# Some Interfaces similar to python in C++

C++:

```
4 4 auto foo(auto a, auto b){
5 5 |     return std::tuple{a, b, a+b};
6 6 |
7 6 }
8 7
9 8 int main(){
10 9 |     auto [_, _, c] = foo(1, 2);
11 10 };
```



# Why Can't We Just Use Python?

# So why can't we just use Python

## ❑ Interpreted Language

- ❑ slow



- ❑ Easier to make mistakes

# Let's Talk About Some Of The Problems



## Problems:

```
1  def foo(a):  
2      a + 1  
3  
4  if __name__ == "__main__":  
5      print(foo(10) + 10)  
6
```

## Problems:

```
1 def foo(a):  
2     return a + b  
3  
4 if __name__ == "__main__":  
5     b = 10  
6     print(foo(10))
```

# Problems:

```
1 ▾ def foo(a, b):  
2     return a + b  
3  
4 ▾ def bar(a, b):  
5     a + [b]  
6  
7 ▾ if __name__ == "__main__":  
8     print(foo(10, 10))  
9
```



# Problems:

```
Traceback (most recent call last):
```

```
File "main.py", line 8, in <module>
```

```
    bar(10, 10)
```

```
File "main.py", line 5, in bar
```

```
    a + [b]
```

```
TypeError: unsupported operand type(s) for +: 'int' and 'list'
```

## Problems:

```
1  what_is_this = 4294967295
2
3  def foo(x, y):
4      4294967296
5
6  if __name__ == "__main__":
7      print(foo(what_is_this, 1))
```


## Problems:

```
1 def foo(a):  
2     return bin(a), type(bin(a))  
  
( '0b10101101100111', <class 'str'>)  
  
5     print(foo(1111))  
6
```




# Some Solutions In Newer Versions

# Some Solutions: Python 12 Typing

```
1 from typing import List
2
3 def foo(a: int, b: int) -> int: 1 usage
4     return a+b
5
6  if __name__ == '__main__':
7     print(foo(a: 1, b: 2.1))
```



Expected type 'int', got 'float' instead

 test


```
def foo(a: int,
        b: int) -> int
```



# Some Solutions: Python 12 Typing

```
3 def foo(a: int, b: int) -> int: 1 usage
4      a+b
5
6  if __name__ == '__main__':
7     print(foo(a: 1, b: 2.1))
```

Expected to return 'int', got no return

 builtins

class int

int([x]) -> integer int(x, base=10) -> integer

Convert a number or string to an integer, or return 0 if no arguments are given. If x is a number, return x.\_\_int\_\_(). For floating point numbers, this truncates towards zero.

If x is not a number or if base is given, then x must be a string, bytes, or bytearray instance representing an integer literal in the given base. The literal can be preceded by '+' or '-' and be surrounded by whitespace. The base defaults to 10. Valid bases are 0 and 2-36. Base 0 means to interpret the base from the string as an integer literal. >>> int('0b100', base=0) 4

[`int` on docs.python.org](#)

# Some Solutions: Python 12 Typing

```
1 from typing import List, Tuple
2
3 def foo[T: (int, str)](a: List[T], b: T) -> List[T]:
4     a.append(b)
5     return a
6
7 ▶ if __name__ == '__main__':
8     print(foo(a=[1, "2"], b=3))
```



# Some Solutions: Python 12 Typing

```
1 from typing import List, Tuple
2
3 def foo[T: (int, str)](a: List[T], b: T) -> List[T]: 1 usage
4     a.append(b)
5     return a
6
7 ▶ if __name__ == '__main__':
8     print(foo(a: [1, "2"], b: 3.1))
9
```

Expected type 'int | str' (matched generic type 'T ≤: int | str'), got 'float' instead

# Some Solutions: Python 12 Typing

```
1 from typing import List
2
3 type U[T] = list[T] | tuple[T, ...]
4
5
6 def foo(a: U) -> List[U]: 2 usages
7     return [a]
8
9 ▶ if __name__ == '__main__':
10     print(foo([20.1]))
11     print(foo((20.1, 20.2, 20.3)))
```

# Some Solutions: Linters

- ❑ Flake 8
- ❑ MyPy

```
1 from typing import List, Tuple
```

```
2
```

```
mypy C:\Users\caleb\PycharmProjects\pythonProject\test.py --enable-incomplete-feature=NewGenericSyntax .
```

```
C:\Users\caleb\PycharmProjects\pythonProject\test.py:4: error: List item 0 has incompatible type "str"; expected "int" [list-item]
```

```
C:\Users\caleb\PycharmProjects\pythonProject\test.py:7: error: Value of type variable "T" of "foo" cannot be "float" [type-var]
```

```
4 return a + ["END!"]
```

```
5
```

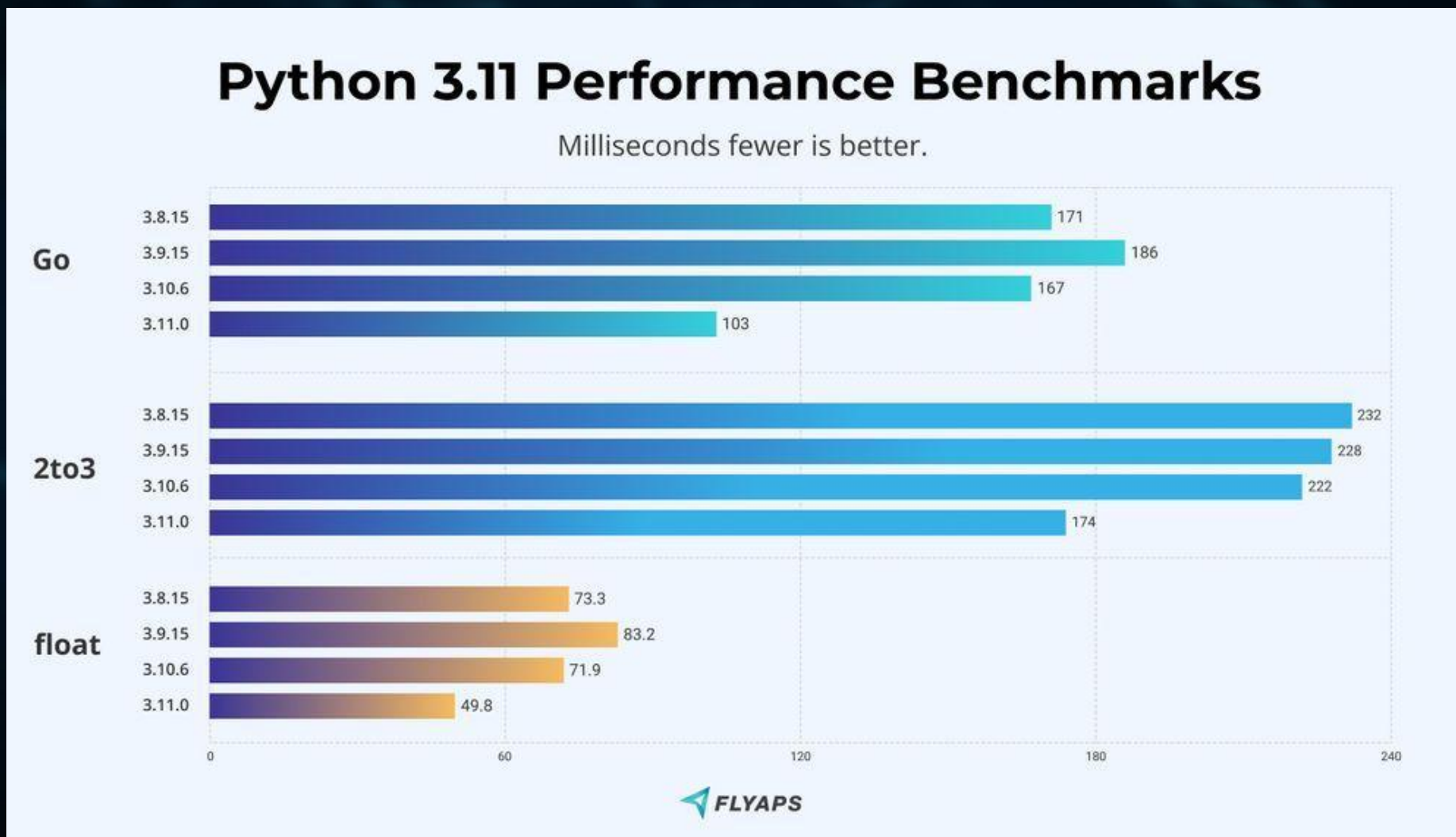
```
6  if __name__ == '__main__':
```

```
7 print(foo([20.1]))
```

# Performance



# Some Solutions: Python is getting faster



# Some Solutions: Importance Of Inner Mechanics

```
1 import timeit
2 from random import uniform
3
4 rand_val = 100000
5 a = [uniform(a:0, rand_val) for i in range(rand_val)]
6
7 def func_1():
8     s = 0
9     for i in a:
10        s += i
11    return s
12
13 def func_2():
14    return sum(a)
15
16 ▶ if __name__ == '__main__':
17    print(timeit.timeit(stmt='func_1()', number=rand_val, globals=globals()))
18    print(timeit.timeit(stmt='func_2()', number=rand_val, globals=globals()))
```

266.0457221000106

64.64083260000916

# Some Solutions: can we get better?!

```
1 import timeit
2 import numpy as np
3
4 rand_val = 100000
5 a = np.random.randn(rand_val)
6                                     3.3721860999939963
7 def func_3():
8     np.sum(a)
9
10 ▶ if __name__ == '__main__':
11     print(timeit.timeit(stmt='func_3()', number=rand_val, globals=globals()))
```

## Some Solutions: what is Numpy

- ❑ Abstraction over contiguous multidimensional array
- ❑ Export easy to use API for many mathematical functions
- ❑ Arrays are not dynamic
- ❑ Core implementation is in C and C++ also all types are bound to the language
- ❑ Usually constructed for basic C++ types



# Some Solutions: Importance of Using Correct Types

```
1 import timeit
2 from random import uniform
3 import numpy as np
4
5 rand_val = 50000
6 a = [uniform(a:0, rand_val) for i in range(rand_val)]
7
8 def func_2():      16.656653799989726
9     return sum(a)  121.12174510001205
10
11 def func_3():
12     return np.sum(a)
13
14 ▶ if __name__ == '__main__':
15     print(timeit.timeit(stmt='func_2()', number=rand_val, globals=globals()))
16     print(timeit.timeit(stmt='func_3()', number=rand_val, globals=globals()))
```

# Some Solutions: Importance of Using Correct Types



## Some Solutions: Why is Pure python better than Numpy

- ❑ We have an overhead
  - ❑ Data was created as a pure python list
  - ❑ We need to copy the whole data to a ndarray
  - ❑ Also data has to be copied from Python memory to C Memory space



## Some Solutions: Takeaways

- ❑ Always use modules that already define the functionality you need  
(no need to write everything by yourself)
- ❑ Learn the Module you will use
  - ❑ Use functionality as expected
  - ❑ Don't abuse the nice API



# More Open Source Modules

# PyTorch

# What is PyTorch

- ❑ Abstraction over contiguous multidimensional array (Sound familiar :))
- ❑ Abstraction over AI models and parts
- ❑ Easy way to mix and match to build new AI modules
- ❑ Core Implementation in C++ AND CUDA
- ❑ Very easy to work with CPU and CUDA
- ❑ Easy way to train new AI models

# Examples: Matrix Calculation on CPU

```
import torch as pt
import timeit

a = pt.randn(1000000, 1000000)
b = pt.arange(5000000).reshape(5, 1000000)
print(timeit.timeit("(a*b)", number=1000, globals=globals()))
```

39.814935477000006



# Examples: Matrix Calculation on GPU

```
import torch as pt
import timeit
device = pt.device('cuda:0')
a = pt.randn(10000000, device=device).reshape(2, 5, 10000000)
b = pt.arange(50000000, device=device).reshape(5, 10000000)
print(timeit.timeit("(a*b)", number=1000, globals=globals()))
```

0.04776061099983053

## Examples: Matrix Calculation on GPU (full round trip)

```
import torch as pt
import timeit
device = pt.device('cuda:0')
a = pt.randn(1000000, 5, 1000000, device=device)
b = pt.arange(5000000, device=device).reshape(5, 1000000)
print(timeit.timeit("(a*b).to('cpu')", number=100, globals=globals()))
```

31.858522263000168

# Examples: Simple Network

```
from torch import nn as ptnn
model = ptnn.Sequential(
    ptnn.Linear(in_features: 784, out_features: 4096), ptnn.ReLU(),
    ptnn.Dropout(),
    ptnn.BatchNorm1d(4096),
    ptnn.Linear(in_features: 4096, out_features: 2048), ptnn.ReLU(),
    ptnn.Dropout(),
    ptnn.BatchNorm1d(2048),
    ptnn.Linear(in_features: 2048, out_features: 1024), ptnn.ReLU(),
    ptnn.Dropout(),
    ptnn.BatchNorm1d(1024),
    ptnn.Linear(in_features: 1024, out_features: 512), ptnn.ReLU(),
    ptnn.Dropout(),
    ptnn.BatchNorm1d(512),
    ptnn.Linear(in_features: 512, out_features: 256), ptnn.ReLU(),
    ptnn.Dropout(),
    ptnn.BatchNorm1d(256),
    ptnn.Linear(in_features: 256, out_features: 10), ptnn.ReLU(),
    ptnn.LogSoftmax(dim=1)).to("cuda:0")
```



# Examples: Simple Network





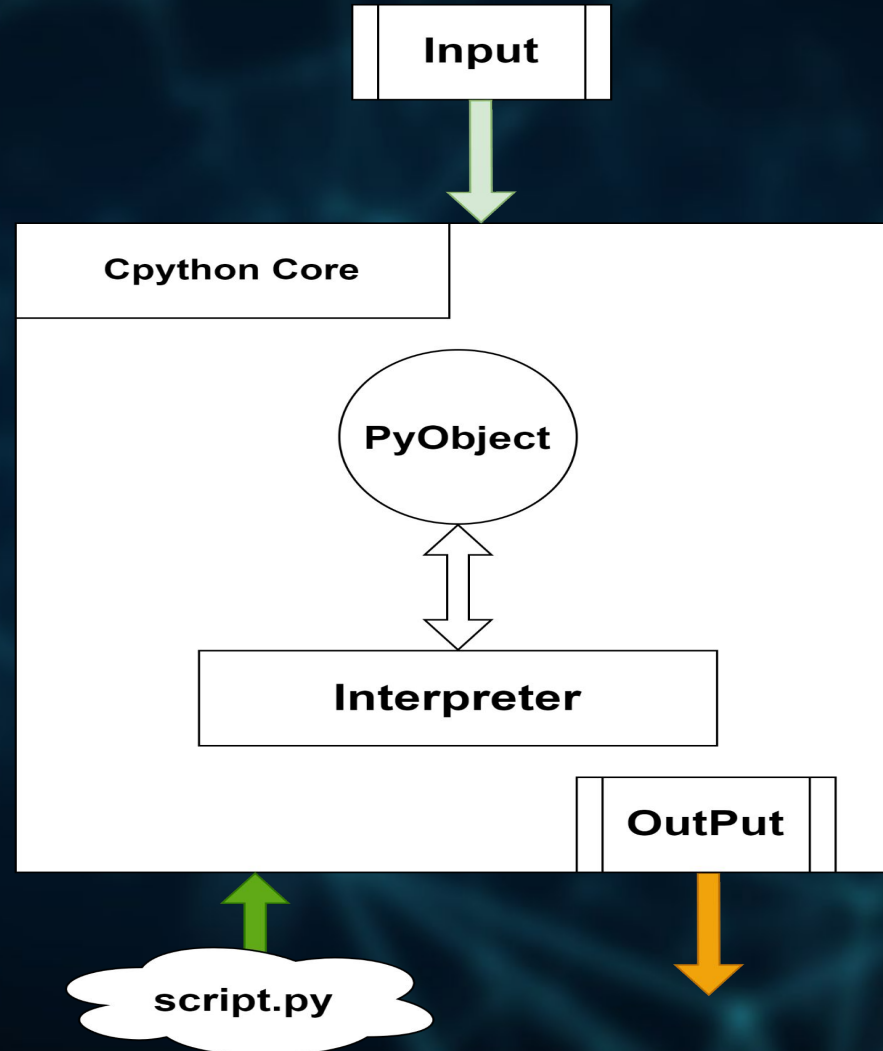
# Making Our Own Binding

## Sometimes we need our own functionality

```
4 float foo(int a, int b){  
5     |   return (2*a + 3*b) * 0.5;  
6     }  
    }
```

- ❑ The Python Interpreter
  - ❑ One of the references and the most used one
  - ❑ its a C program that interprets the python programming language and runs it
  - ❑ Maintains the internal state of the python program
  - ❑ Everything is a PyObject

# CPython

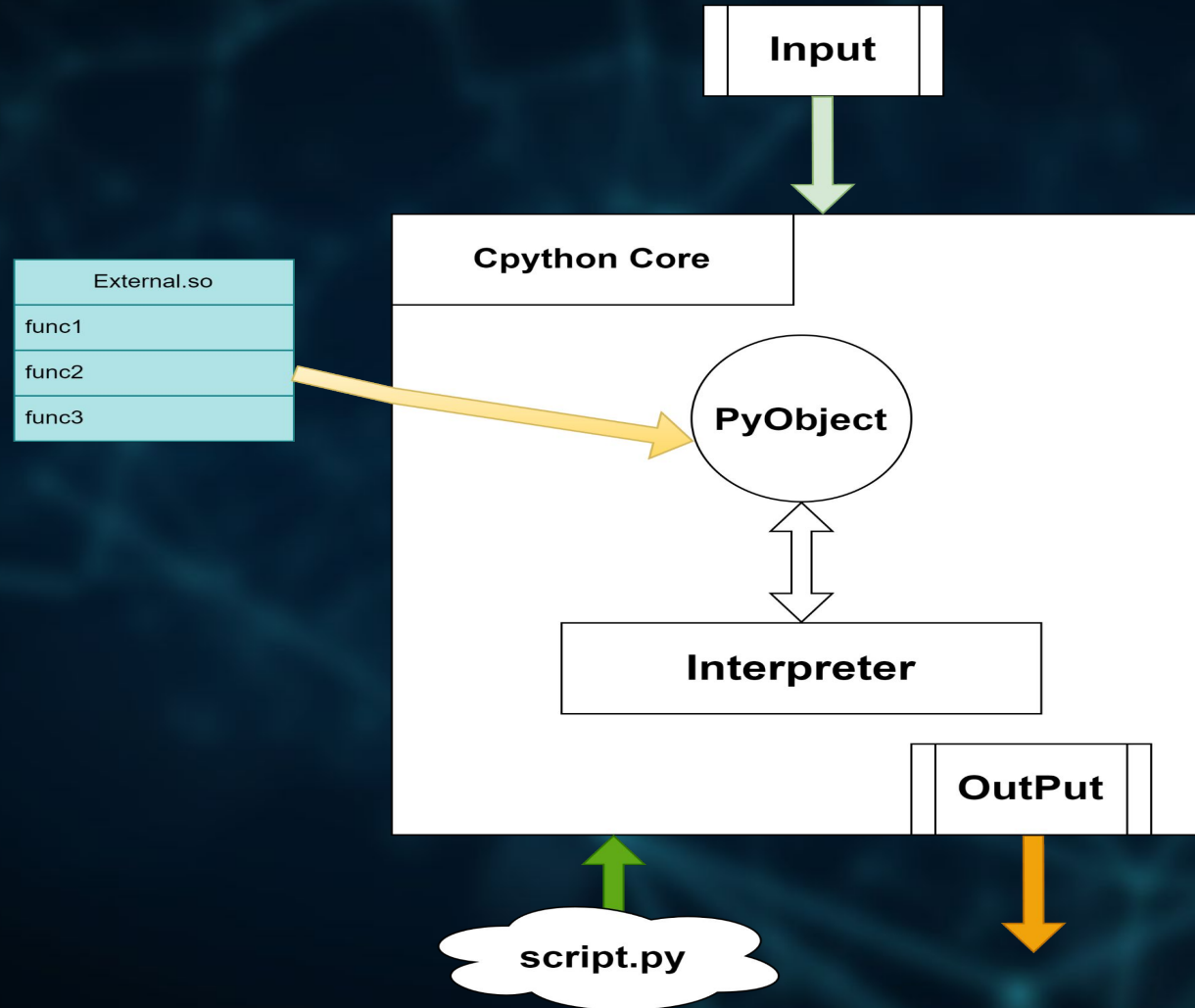




# CPython: Extension Model

- ❑ CPython allows us to extend the language
  - ❑ Provides Python.h file that contains the API
  - ❑ Allows us to add hooks to be called from python

# CPython



# CPython: Extension Model

```
#include <Python.h>

// C function
float foo(int a, int b) {
    return (2 * a + 3 * b) * 0.5;
}
```

# CPython: Extension Model

```
// CPython wrapper for foo
static PyObject* py_foo(PyObject* self, PyObject* args) {
    int a, b;
    // Parse the Python arguments (expecting two integers)
    if (!PyArg_ParseTuple(args, "ii", &a, &b)) {
        return NULL; // If parsing fails, return NULL
    }

    // Call the actual C function
    float result = foo(a, b);

    // Return the result as a Python float object
    return PyFloat_FromDouble(result);
}
```



# CPython: Extension Model

```
// Method definitions
static PyMethodDef FooMethods[] = {
    {"foo", py_foo, METH_VARARGS, "Calculate (2*a + 3*b) * 0.5"},
    {NULL, NULL, 0, NULL} // Sentinel
};
```

# CPython: Extension Model

```
// Module definition
static struct PyModuleDef foomodule = {
    PyModuleDef_HEAD_INIT,
    "foo_module", // Name of the module
    "A module that provides foo()", // Module documentation
    -1, // Size of per-interpreter state or -1 if global
    FooMethods // Methods in the module
};
```

# CPython: Extension Model

```
// Module initialization function
PyMODINIT_FUNC PyInit_foo_module(void) {
    return PyModule_Create(&foomodule);
}
```

# CPython: Extension Model

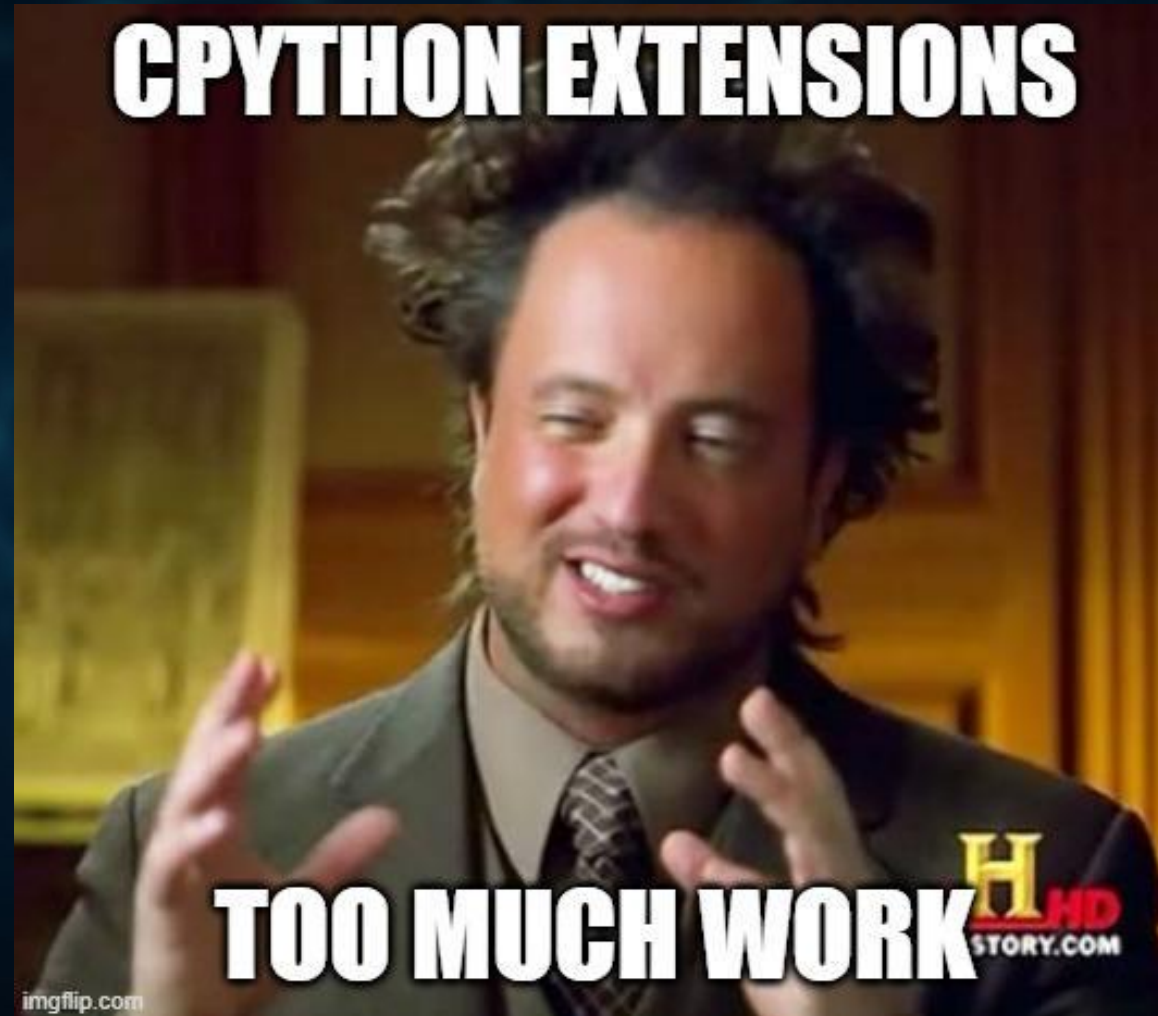
```
from setuptools import setup, Extension

# Define the extension module
module = Extension("foo_module", sources=["foo_module.c"])

# Setup function
setup(
    name="foo_module",
    version="1.0",
    description="Python interface for the foo C function",
    ext_modules=[module],
)
```



# CPython: Extension Model



# Binding Tools

# Binding Tools

There are many Binding tools , during this talk we would focus on

- ❑ Static Binding from C++ to Python
  - ❑ PyBind11
  - ❑ Boost Python
- ❑ Dynamic Binding from Cpp to Python
  - ❑ CPPyy
- ❑ Static Binding language from C++ to Python and from Python to C++
  - ❑ Cython

# Binding Tools: PyBind11

```
1  #include <pybind11/pybind11.h>
2
3  namespace py = pybind11;
4
5  // The C++ function
6  float foo(int a, int b) {
7      return (2 * a + 3 * b) * 0.5;
8  }
9
10 // Module definition
11 PYBIND11_MODULE(foo_module, m) {
12     // Optional module docstring
13     m.doc() = "A module that provides the foo function";
14
15     // Expose the foo function
16     m.def("foo", &foo, "A function that calculates (2*a + 3*b) * 0.5",
17         // Optional argument names for better Python usability
18         py::arg("a"), py::arg("b"));
19 }
```



# Binding Tools: PyBind11

```
3  cmake_minimum_required(VERSION 3.29)
4  project(Bindings)
5
6  set(CMAKE_CXX_STANDARD 26)
7  find_package(pybind11 REQUIRED)
8
9  add_library(foomod SHARED main.cpp)
10 target_link_libraries(foomod pybind11_all_do_not_use)
```

# Binding Tools: PyBind11

```
1  from setuptools import setup
2
3  setup(
4      name="foo_module",
5      version="1.0",
6      description="Python interface for the foo C++ function using pybind11",
7  )
```

# Binding Tools: PyBind11

```
L2  install(CODE "execute_process(  
L3      WORKING_DIRECTORY ${CMAKE_BINARY_DIR}  
L4      COMMAND ${Python_EXECUTABLE}  
L5      ${CMAKE_CURRENT_LIST_DIR}/setup.py install")
```

# Binding Tools: Boost Python

```
1 #include <boost/python.hpp>
2
3 // C++ function
4 float foo(int a, int b) {
5     return (2 * a + 3 * b) * 0.5;
6 }
7
8 // Module definition
9 BOOST_PYTHON_MODULE(foo_module) {
10     using bp = boost::python;
11
12     // Expose the C++ function 'foo' to Python
13     bp.def("foo", foo, "A function that calculates (2*a + 3*b) * 0.5");
14 }
```



# Binding Tools: Boost Python

```
3 cmake_minimum_required(VERSION 3.29)
4 project(Bindings)
5
6 set(CMAKE_CXX_STANDARD 26)
7 find_package(Boost COMPONENTS python3 REQUIRED)
8 find_package(PythonInterp 3 REQUIRED)
9 find_package(PythonLibs 3 REQUIRED)
10
11
12 add_library(foomod SHARED main.cpp)
13 target_link_libraries(foomod Boost::python3)
```

# Binding Tools: What about classes?

```
// C++ class definition
struct cpp_class {
    float foo(int a, int b) {
        return (2 * a + 3 * b) * 0.5;
    }

    float bar(int a, int b) {
        return (a + b) * 0.5;
    }
};
```

# Binding Tools: What about classes?

```
16 // Exposing cpp_class to Python using pybind11
17 PYBIND11_MODULE(cpp_class_module, m) {
18     py::class_<cpp_class>(m, "cpp_class")
19         .def(py::init<>()) // Expose the default constructor
20         .def("foo", &cpp_class::foo, "A function that calculates (2*a + 3*b) * 0.5")
21         .def("bar", &cpp_class::bar, "A function that calculates (a + b) * 0.5");
22 }
```

# What About Performance



# Binding Packs: But (Performance)

Platform	Time (second) [10 million runs]
C++	0.0042 (4.24 mili)
Python	1.2206 (1220.60 mili)
CPython	1.1058 (1105.80 mili)
PyBind11	5.74263 (5742.63 mili)

# Binding Packs: But (Performance)

We have overheads

- ❑ The data needs to go back and forth from C++ object to PyObject
- ❑ Different Libraries use different methods
  - ❑ PyBind11 allocates almost everything as smart pointers or vectors and it makes it a bit slower
- ❑ Just like with CUDA we need to select the right parts to be optimised.

# Binding For Better Performance

- ❑ Make Batch functions (Hot areas of code)
- ❑ Use Bindings objects for C++ conversion (`py::dict`)
- ❑ Use built in optimised type bindings like (`py::array_t<double>`)
- ❑ Return values: important to understand the lifetime of the returned object
- ❑ Use smart pointers instead of regular pointers

# Binding Packs: Better Function for python run

```
✓ float foo(py::array_t<int> a) {  
    float sum = 0;  
    float v = 2;  
    const auto ptr = static_cast<const int*>(a.request().ptr);  
    for (int i = 0; i < a.size(); i++) {  
        sum += v++ * static_cast<float>(ptr[i]);  
    }  
}
```



# Binding Packs: But (Performance)

Platform	Time (second) [10 million runs, 15 values]
C++	0.0624 (62.4 mili)
PyBind11	1.64263 (1642.63 mili)

# Cython

# Cython

- ❑ PyBind11 and BoostPython are C++ language extensions
  - ❑ Can run only C++ code in python
- ❑ Cython
  - ❑ Rich language that looks like python
  - ❑ More features and much more flexible
  - ❑ We can run Python code from C++

# C++ -> Python



# Cython: It All Starts from CPP class

```
namespace important{
    struct abstract{
        virtual float foo(int a, int b) = 0;
        virtual float bar(int a, int b) = 0;
        virtual ~abstract() = default;
    };

    struct real : public abstract{
        float foo(int a, int b) override {
            return (2*a+3*b)*0.5;
        }

        float bar(int a, int b) override {
            return (a+b)*0.5;
        }
    };
}
```

# Cython: PXD File

```
2 for libcpp.memory cimport shared_ptr
3
4 cdef extern from "class.hpp" namespace "important":
5     cdef cppclass Cabstract "important::abstract":
6         abstract() except+
7         float foo(int a, int b)
8         float bar(int a, int b)
9
10
11     cdef cppclass Creal(Cabstract):
12         real() except+
13         float foo(int a, int b)
14         float bar(int a, int b)
```

# Cython: PXD File

```
cdef class PyReal:  
    cdef shared_ptr[CReal] c_real
```

# Cython: PYX File

```
from my_module cimport *
from libcpp.memory cimport make_shared, nullptr
from cython.operator cimport dereference

cdef class PyReal:
    def __cinit__(self):
        c_real = make_shared[CReal]();

    def __init__(self):
        pass

    cpdef foo(int a, int b):
        dereference(c_real.get()).foo(a, b)
```



# Cython: Cythonization

```
cython -3 --cpplus my_model.pyx
```

# Cython: Cythonization Cmake Version

```
1  cmake_minimum_required(VERSION 3.29)
2  project(Bindings)
3
4  find_package(Python 3 REQUIRED COMPONENTS Interpreter Development)
5
6  add_custom_command(
7      OUTPUT
8          ${CMAKE_CURRENT_BINARY_DIR}/my_model.cpp
9      DEPENDS
10         ${CMAKE_CURRENT_BINARY_DIR}/my_model.pyx
11         ${CMAKE_CURRENT_BINARY_DIR}/my_model.pxd
12
13     COMMAND cython -3 --cplus ${CMAKE_CURRENT_BINARY_DIR}/my_model.pyx
14 )
15
16 add_library(my_model SHARED my_model.cpp)
17
18 #we want it to be without lib....
19 set_target_properties(py_my_modeltypes PROPERTIES PREFIX "")
20
21 python_install_function(...)
```

# Python -> C++

# Cython: Public Functions

```
cdef public string load_json(string path):  
    j = json.load(path)  
    j['x'] = 'PYTHON'  
    return j.dumps()
```



# Cython: C++ Init Python

```
3 void PythonAutoLoader() {
4     const char* libpython_path = "python so file";
5     python_lib_handle_ = dlopen(libpython_path, RTLD_NOW | RTLD_GLOBAL);
6     if (nullptr == python_lib_handle_) {
7         throw std::runtime_error("Cannot load libpython");
8     }
9
10    PyImport_AppendInittab("my_model", PyInit_my_model);
11    Py_Initialize();
12    if (!PyImport_ImportModule("my_model")) {
13        PyErr_Print();
14        throw std::runtime_error("Cannot import my_model python interop module");
15    }
16 }
```

# Few Words About Exceptions

# Cython: Exceptions

- ❑ We cannot propagate exceptions from python to C++
- ❑ It's easy to create a function in C++ that is invoked from cython

```
2 cdef extern from "Thrower.h" namespace "utils":  
3     cdef void ThrowCppRuntimeError(string message)  
4  
5 def throw_cpp_error(msg: string):  
6     ThrowCppRuntimeError(msg.encode("utf8"))
```

# Dynamic Binding



# Dynamic Binding: Cppyy

- ❑ PyBind11, BoostPython and Cython are static
  - ❑ need to be compiled into a module
- ❑ Cppyy is dynamic binding
  - ❑ write C++ directly into python
  - ❑ supports CPython and PyPy (much faster on PyPy)
- ❑ Easy to use functionality

# Dynamic Binding: Cppyy

```
cppyy.cppdef(r"""\  
    float foo(int a, int b) {  
        return (2*a+3*b)*0.5; } """)  
cppyy.gbl.foo(10, 20)  
40.0
```

# Dynamic Binding: Cppyy

Platform	Time (second) [10 million runs, 15 values]
C++	0.0042 (4.24 mili)
Cppyy	0.7243 (724.3 mili)

# Dynamic Binding: Cppyy

- ❑ Uses Cling/ clang base interpreter for C++
  - ❑ No pre install or compile
  - ❑ No need to duplicate STL
  - ❑ Full support for templates
  - ❑ Full support of inheritance
  - ❑ Full support for callbacks and lambdas



# Cppy : Examples

```
struct abstract {
    virtual int foo(int a, int b) = 0;
    virtual std::array<int, 10> bar(int n) = 0;
    virtual ~abstract() = default;
};

struct real : public abstract {
    real() = default;

    int foo(int a, int b) override {
        return static_cast<int>((2*a + 3*b)*0.5);
    }

    std::array<int, 10> bar(int n) override {
        return std::array{1*n, 2, 3, 4, 5, 6, 7, 8, 9, 10*n};
    }
};
```

## Cppy : Examples (including)

```
>>> import cppy
>>> cppy.include('features.h')
True
>>> from cppy.gbl import real
>>> real
<class cppy.gbl.real at 0x40d2b00>
```

# Cppy : Examples (Calling C++ Class Method)

```
>>> r = real()  
>>> r.foo(10, 20)
```

# Cppy : Examples (Cross Inheritance)

```
>>> cppy.include('array')
True
>>> class PyReal(cppy.gbl.abstract):
...     def foo(self, a, b):
...         return a+b+10
...     def bar(self, a):
...         return cppy.gbl.array[int, 10]()
...
>>> pr = PyReal()
>>> pr.foo(10, 20)
40
```



# Cppy : Examples (Template Functions)

```
template <typename R, typename... U, typename... Args>  
R callit(R(*f)(U...), Args&&... args) {  
    return std::invoke(f, std::forward<Args>(args)...);  
}
```

## Cppy : Examples (Template Functions)

```
>>> from cppy.gbl import callit
>>> def f(a: int, b: int) -> float:
...     return (2*a+3*b)*0.5
...
>>> callit(f, 10, 20)
40.0
```

# Closing Notes

# Dynamic Binding: Cppyy

- ❑ Use the right tools for your needs
- ❑ Don't implement everything by yourself
  - ❑ Be “lazy” and use open source
  - ❑ Try cythonizing pure python code to get performance
- ❑ If you have to write something by yourself
  - ❑ Be familiar with the rules of the binding pack your using
  - ❑ Understand how the transfer of data works
  - ❑ Avoid dangling pointers
  - ❑ Be aware of Threading problems in Python



# QUESTIONS



# THANK YOU FOR LISTENING

**ALEX DATHSKOVSKY**

[ALEX.DATHSKOVSKY@SPEEDATA.IO](mailto:ALEX.DATHSKOVSKY@SPEEDATA.IO)

[WWW.LINKEDIN.COM/IN/ALEXDATHSKOVSKY](http://WWW.LINKEDIN.COM/IN/ALEXDATHSKOVSKY)